

# WEAK CONVERGENCE OF THE SIMPLE LINEAR RANK STATISTIC UNDER MIXING CONDITIONS IN THE NONSTATIONARY CASE\*

M. HAREL<sup>†</sup> AND M. L. PURI<sup>†‡</sup>

**Abstract.** The asymptotic distribution theory of simple linear rank statistics for the case when the underlying random variables are nonstationary is studied both for the  $\varphi$ -mixing and strong mixing cases.

**Key words.** weak convergence; linear rank statistics, empirical processes,  $\varphi$ -mixing, strong mixing

**1. Introduction.** Let  $X_{ni}$ ,  $1 \leq i \leq n$ ,  $n \geq 1$ , be real-valued r.v.'s (random variables) with continuous d.f.'s (distribution functions)  $F_{ni}(x)$ ,  $x \in \mathbf{R}$ , and let  $c_{ni}$  ( $1 \leq i \leq n$ ,  $n \geq 1$ ) be an array of regression constants defined by a function  $g$  on  $[0, 1]$  as

$$(1.1) \quad c_{ni} = g(i/n), \quad 1 \leq i \leq n, \quad n \geq 1.$$

Denote by  $\hat{H}_n(x) = n^{-1} \sum_{i=1}^n c_{ni} \mathbf{1}_{[X_{ni} \leq x]}$  the weighted empirical process where  $\mathbf{1}_{[\cdot]}$  denotes the indicator function and by  $\hat{F}_n(x) = n^{-1} \sum_{i=1}^n \mathbf{1}_{[X_{ni} \leq x]}$  the usual empirical process.

The corresponding expectations are denoted by

$$H_n = \mathbf{E}(\hat{H}_n) \quad \text{and} \quad F_n = \mathbf{E}(\hat{F}_n).$$

We will study the asymptotic behavior of the simple linear rank statistic of the form

$$(1.2) \quad \mathcal{S}_n(J) = n^{1/2} \int_{-\infty}^{+\infty} J\left(\frac{n}{n+1} \hat{F}_n(x)\right) d\hat{H}_n(x) \int_{-\infty}^{+\infty} J(F_n(x)) dH_n(x),$$

where  $J$  is a score function defined on the open unit interval.

The problem of finding a sufficiently large class of score functions for which the linear rank statistic is asymptotically normal was first considered by Chernoff and Savage [3]. Their results were later on strengthened considerably by several authors, mainly by Govindarajulu et al. [8], Pyke and Shorack [10], Hájek [9] and Dupač and Hájek [6] for the independent case, by Fears and Mehra [7] for the  $\varphi$ -mixing case with stationary random variables, and by Denker and Rösler [4] for the  $\varphi$ -mixing as well as strong mixing case but under a *stationary* set-up. In this paper we investigate the asymptotic distribution theory of the simple linear rank statistics (1.2) for the case when the underlying random variables are *nonstationary*.

**2. Preliminaries.** In this section, we give some propositions that are minor variations of Denker and Rösler [4] and so their proofs will be either omitted or briefly outlined.

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<sup>†</sup>Department of Mathematics, Indiana University, Bloomington, Indiana 47405.

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For  $\delta \geq 0$ , set  $\eta = \delta(4 + 2\delta)^{-1}$ .

Let  $\mu_\delta$  denote the measure on  $[0, 1]$  given by its density  $(z(1 - z))^{-1/2-\eta}$  relative to the Lebesgue measure. For a monotone function  $J$ , let  $\|J\|_\delta$  be the  $\mathbf{L}_1$ -norm of  $J$  in  $\mathbf{L}_1(\mu_\delta)$ . By the Jordan decomposition of any right continuous function,  $J$  has a unique decomposition:  $J = J_1 - J_2$  where  $J_1$  and  $J_2$  are monotone functions and  $J_1(\frac{1}{2}) = 0$ . For such a function, we set

$$\|J\|_\delta = \|J_1\|_\delta + \|J_2\|_\delta,$$

where  $J_1$  and  $J_2$  belong to  $\mathbf{L}_1(\mu_\delta)$ .

Denote by  $\mathcal{H}_\delta$  the space of all right continuous functions  $J$  with  $\|J\|_\delta < \infty$  and  $J(\frac{1}{2}) = 0$  and let  $\mathcal{G}_\delta$  be the set of all  $J \in \mathcal{H}_\delta$  for which the measure  $\nu$  defined by  $J = \int d\nu$  is absolutely continuous with respect to Lebesgue measure. It is well known that  $\mathcal{G}_\delta$  is the  $\|\cdot\|_\delta$ -norm closure of  $\mathbf{C}_{2,b}$ : the space of functions with bounded second derivatives. The a priori assumption of having the space  $\mathcal{H}_\delta$  of right continuous functions  $J$  with  $J(\frac{1}{2}) = 0$  is no restriction because  $\bar{J}$  defined by  $\bar{J}(x) = J(x+) = \lim_{y \downarrow x} J(y)$  is a well-defined right continuous function and if  $\bar{J}$  belongs to  $\mathcal{H}_\delta$ ,  $\mathcal{S}_n(J)$  and  $\mathcal{S}_n(\bar{J})$  are asymptotically equivalent (see [4]).

We consider the array  $G_{ni}$  ( $1 \leq i \leq n$ ,  $n \geq 1$ ) of d.f.'s on  $[0, 1]$  defined by

$$(2.1) \quad G_{ni} = F_{ni} \circ F_n^{-1}.$$

Denote by  $\hat{G}_n$  the empirical process in  $[0, 1]$  derived from  $\hat{H}_n$  and defined by

$$(2.2) \quad \hat{G}_n(t) = n^{-1} \sum_{i=1}^n c_{ni} \mathbf{1}_{[F_n(X_{ni}) \leq t]}, \quad t \in [0, 1],$$

and  $\hat{I}_n$  the empirical process on  $[0, 1]$  derived from  $\hat{F}_n$  and defined by

$$(2.3) \quad \hat{I}_n(t) = n^{-1} \sum_{i=1}^n \mathbf{1}_{[F_n(X_{ni}) \leq t]}, \quad t \in [0, 1].$$

We also denote  $G_n = \mathbf{E}(\hat{G}_n)$  and  $I_n = \mathbf{E}(\hat{I}_n)$ . The linear rank statistic defined in (1.2) can then be written as

$$(2.4) \quad \mathcal{S}_n(J) = n^{1/2} \int_0^1 J\left(\frac{n}{n+1} \hat{I}_n(t)\right) d\hat{G}_n(t) - n^{1/2} \int_0^1 J(I_n(t)) dG_n(t).$$

The connection between the dependence structure of the processes and the class of functions for which asymptotic normality holds is expressed by the following condition:

$$(2.5) \quad \begin{cases} n \mathbf{E}(\hat{G}_n(t) - G_n(t))^2 \leq C \Lambda_n^2(t(1-t))^{1-2\eta}, \\ n \mathbf{E}(\hat{I}_n(t) - I_n(t))^2 \leq C(t(1-t))^{1-2\eta}, \end{cases}$$

for all  $t \in (0, 1)$  and  $n \geq 1$  where  $\eta = \delta(4 + 2\delta)^{-1}$ ,  $\Lambda_n = \sup_{1 \leq i \leq n} |g(i/n)|$  for  $g$  defined in (1.1) and  $C$  is some positive constant.

**PROPOSITION 2.1.** *Let  $K > 0$  and  $\delta \geq 0$  be given. Then there exists a constant  $C_1$  such that the following holds: if  $\{X_{ni}\}$  is an array of r.v.'s satisfying (2.5) and if  $\{c_{ni}\}$  are regression constants defined by (1.1), we have*

$$n \mathbf{E} \left\{ \int_{-\infty}^{+\infty} J\left(\frac{n}{n+1} \hat{F}_n(x)\right) d\hat{H}_n(x) \right.$$

$$(2.6) \quad - \int_{-\infty}^{+\infty} J\left(\frac{n}{n+1} F_n(x)\right) d\widehat{H}_n(x) \Big\}^2 \leq C_1 \Lambda_n^2 \|J\|_\delta^2,$$

$$(2.7) \quad n \mathbf{E} \left\{ \int_{-\infty}^{+\infty} J\left(\frac{n}{n+1} F_n(x)\right) d(\widehat{H}_n - H_n)(x) \right\}^2 \leq C_1 \Lambda_n^2 \|J\|_\delta^2,$$

$$(2.8) \quad n \left\{ \int_{-\infty}^{+\infty} \left( J(F_n(x)) - J\left(\frac{n}{n+1} F_n(x)\right) \right) dH_n(x) \right\}^2 \leq C_1 \Lambda_n^2 \|J\|_\delta^2,$$

$$(2.9) \quad n \mathbf{E} (\mathcal{S}_n(J))^2 \leq C_1 \Lambda_n^2 \|J\|_\delta^2.$$

*Proof.* We only prove (2.6) because our method of proof is similar to that of Proposition 2 in [4]. It suffices to consider an increasing function  $J \in \mathcal{H}_\delta$ . Define

$$\varphi(x, t) = \begin{cases} 1 & \text{if } \frac{n}{n+1} F_n(x) \leq t < \frac{n}{n+1} \widehat{F}_n(x), \\ -1 & \text{if } \frac{n}{n+1} \widehat{F}_n(x) \leq t < \frac{n}{n+1} F_n(x), \\ 0 & \text{otherwise} \end{cases}$$

and denote by  $\widehat{F}_n^{-1}(t) = \inf\{x \in \mathbf{R}: \widehat{F}_n(x) \geq t\}$  the left continuous inverse of  $\widehat{F}_n$ . Since for  $t \leq n/(n+1)$ ,

$$\varphi(x, t) = \begin{cases} 1 & \text{if } \widehat{F}_n^{-1}\left(\frac{n+1}{n} t\right) \leq x < F_n^{-1}\left(\frac{n+1}{n} t\right), \\ -1 & \text{if } F_n^{-1}\left(\frac{n+1}{n} t\right) \leq x < \widehat{F}_n^{-1}\left(\frac{n+1}{n} t\right), \\ 0 & \text{otherwise} \end{cases}$$

it follows that for fixed  $t \leq n/(n+1)$ ,

$$\begin{aligned} \left| \int_{-\infty}^{\infty} \varphi(x, t) d\widehat{H}_n(x) \right| &\leq \Lambda_n \int_{-\infty}^{\infty} |\varphi(x, t)| d\widehat{F}_n(x) \\ &= \Lambda_n \left| \left\langle \frac{n+1}{n} t \right\rangle - \widehat{F}_n \left( F_n^{-1} \left( \frac{n+1}{n} t \right) - \right) \right| \\ &\leq \Lambda_n \left( \left| \widehat{F}_n \left( F_n^{-1} \left( \frac{n+1}{n} t \right) - \right) - \frac{n+1}{n} t \right| + 2 \inf \left\{ 1/n, t \right\} \right) \end{aligned}$$

where  $\langle t \rangle = (k-1)/n$  if  $(k-1)/n < t \leq k/n$ .

From the assumption (2.5) on the sequence  $\{X_{ni}\}$ , it follows for  $t \leq n/(n+1)$  that

$$\begin{aligned} \left( \int_{-\infty}^{+\infty} \varphi(x, t) d\widehat{H}_n(x) \right)^2 &\leq 8\Lambda_n^2 \left( \frac{1}{n} (1 \wedge nt) \right)^2 + 2 \mathbf{E} \left( \widehat{I}_n \left( \frac{n+1}{n} t \right) - I_n \left( \frac{n+1}{n} t \right) \right)^2 \\ &\leq 8\Lambda_n^2 \left( \frac{1}{n} (1 \wedge nt) \right)^2 + 2 \frac{C}{n} \Lambda_n^2 (t(1-t))^{1-2\eta}. \end{aligned}$$

Finally, interchanging the order of integration, and using the Cauchy–Schwarz inequality,

ity, we obtain

$$\begin{aligned}
 & n \mathbf{E} \left( \int_{-\infty}^{+\infty} J \left( \frac{n}{n+1} \widehat{F}_n(x) \right) d\widehat{H}_n(x) - \int_{-\infty}^{+\infty} J \left( \frac{n}{n+1} F_n(x) \right) d\widehat{H}_n(x) \right)^2 \\
 &= n \mathbf{E} \left( \int_0^{n/(n+1)} \int_{-\infty}^{+\infty} \varphi(x, t) d\widehat{H}_n(x) dJ(t) \right)^2 \\
 &\leq n \left( \int_0^{n/(n+1)} \left( \mathbf{E} \left( \int_{-\infty}^{+\infty} \varphi(x, t) d\widehat{H}(x) \right)^2 \right)^{1/2} dJ(t) \right)^2 \\
 &\leq n C_1 \Lambda_n^2 \left( \int_0^{n/(n+1)} \frac{1}{n} (1 \wedge nt) dJ(t) \right. \\
 &\quad \left. + \int_0^{n/(n+1)} n^{-1/2} (t(1-t))^{1/2-\eta} dJ(t) \right)^2 \leq C_1 \Lambda_n^2 \|J\|_\delta^2
 \end{aligned}$$

where  $C_1$  is some positive constant, since

$$n^{1/2} \int_0^{n/(n+1)} (1/n) (1 \wedge nt) dJ(t) \leq C_1 \int_0^1 (t(1-t))^{1/2} dJ(t).$$

The inequality (2.6) is proved.

**PROPOSITION 2.2.** *Let  $\{X_{ni}\}$  satisfy condition (2.5) for some  $\delta > 0$ , and let the regression constants  $c_{ni}$  satisfy (1.1) and  $\sup_{n \in \mathbf{N}} \Lambda_n < +\infty$ . Assume that  $K \subset \mathcal{H}_\delta$  is a subset possessing the following property: for every  $J \in K$ , there exists a normal distribution  $\mathcal{N}(0, \sigma^2)$  where  $0 < \sigma < +\infty$  such that  $S_n(J)$  converges in law to  $\mathcal{N}(0, \sigma^2)$ , then the  $\|\cdot\|_\delta$ -norm closure of  $K$  has the same property.*

*Proof.* Let  $J_1 \in \overline{K}$  where  $\overline{K}$  is the closure of  $K$ , and  $J \in K$ . By Proposition 2.1 and the fact that  $S_n(J_1) = S_n(J) + S_n(J_1 - J)$ , the distributions of  $S_n(J_1)$  and  $S_n(J)$  are closed, uniformly in  $n$ , in the weak topology for sufficiently small  $\|J - J_1\|_\delta$ . This proves the proposition.

For two probability measures  $\mathbf{P}$  and  $\mathbf{Q}$  on  $\mathbf{R}$ , denote by  $D_2(\mathbf{P}, \mathbf{Q}) = \inf(\mathbf{E}(X - Y)^2)^{1/2}$  where the infimum extends over all r.v.'s  $X$  and  $Y$  defined on the same probability space and having distributions  $\mathbf{P}$  and  $\mathbf{Q}$ , respectively.

Let  $\mathcal{L}(Z)$  denote the distribution of the r.v.  $Z$ .

**PROPOSITION 2.3.** *Let  $\{X_{ni}\}$  satisfy condition (2.5) for some  $\delta > 0$  and let  $c_{ni}$  satisfy (1.1) and  $\sup_{n \in \mathbf{N}^*} \Lambda_n < +\infty$ . Assume that there exists an operator  $\sigma: \mathcal{H}_\delta \rightarrow \mathbf{R}$  which is uniformly bounded and satisfies the Lipschitz condition for the  $\|\cdot\|_\delta$  norm.*

*If for every  $J \in K \subset \mathcal{H}_\delta$ ,  $S_n(J)$  converges in law to a normal distribution  $\mathcal{N}(0, \sigma^2(J))$ , then the  $\|\cdot\|_\delta$ -norm closure of  $K$  has the same property.*

*Proof.* Let  $J_1 \in \overline{K}$  and  $J \in K$ . Then, we have

$$\begin{aligned}
 & D_2(\mathcal{L}(S_n(J_1)), \mathcal{N}(0, \sigma^2(J_1))) \leq D_2(\mathcal{L}(S_n(J_1)), \mathcal{L}(S_n(J))) \\
 & \quad + D_2(\mathcal{L}(S_n(J)), \mathcal{N}(0, \sigma^2(J))) + D_2(\mathcal{N}(0, \sigma^2(J)), \mathcal{N}(0, \sigma^2(J_1))) \\
 & \leq D_2(\mathcal{L}(S_n(J)), \mathcal{N}(0, \sigma^2(J))) + C' \|J - J_1\|_\delta
 \end{aligned}$$

for some  $C' > 0$  using Proposition 2.1 and the Lipschitz condition on  $\sigma$ .

Since the convergence in law from  $S_n(J)$  to  $\mathcal{N}(0, \sigma^2(J))$  implies (see [4])

$$\lim_{n \rightarrow \infty} D_2(\mathcal{L}(S_n(J)), \mathcal{N}(0, \sigma^2(J_1))) = 0,$$

the theorem follows.

**3. Convergence of the linear rank statistic.** Recall that the sequence  $\{X_{ni}\}$  is  $\varphi$ -mixing if

$$\sup_{n \geq 1} \sup_{1 \leq j \leq n-m} \left\{ \sup \left\{ \left| \mathbf{P}(B|A) - \mathbf{P}(B) \right|; A \in \sigma(X_{ni}, 1 \leq i \leq j), \right. \right. \\ \left. \left. B \in \sigma(X_{ni}, i \geq j+m) \right\} \right\} = \varphi(m) \downarrow 0$$

for positive integers  $j$  and  $m$ .

Here  $\sigma(X_{n1}, \dots, X_{nj})$  and  $\sigma(X_{n,j+m}, \dots, X_{nn})$  are the  $\sigma$ -fields generated by  $(X_{n1}, \dots, X_{nj})$  and  $(X_{n,j+m}, \dots, X_{nn})$ , respectively.

Also recall that  $\{X_{ni}\}$  satisfies the strong mixing condition if

$$\sup_{n \geq 1} \sup_{1 \leq j \leq n-m} \left\{ \sup \left\{ \left| \mathbf{P}(A \cap B) - \mathbf{P}(A) \mathbf{P}(B) \right|; A \in \sigma(X_{ni}, 1 \leq i \leq j), \right. \right. \\ \left. \left. B \in \sigma(X_{ni}, i \geq j+m) \right\} \right\} = \alpha(m) \downarrow 0.$$

Since  $\alpha(m) \leq \varphi(m)$ , it follows that if  $\{X_{ni}\}$  is  $\varphi$ -mixing, then it is also strong mixing.

We will study the asymptotic behavior of  $\mathcal{S}_n(J)$  when the r.v.'s  $\{X_{ni}\}$  are  $\varphi$ -mixing with rates

$$(3.1) \quad \sum_{m=1}^{+\infty} m(\varphi(m))^{(2+3\delta)/(4+2\delta)} < +\infty \quad \text{for some } 0 \leq \delta < 2$$

or strong mixing with rates

$$(3.2) \quad \sum_{m=1}^{+\infty} m^2 \alpha(m)^{\delta/(2+\delta)} < +\infty \quad \text{for some } \delta > 0.$$

Let  $F_{n,i,j}$  be the d.f. of  $(X_{ni}, X_{nj})$ ,  $1 \leq i \leq j \leq n$ ,  $n \geq 1$ . For any sequence of d.f.'s  $\{G_l^*, l \geq 1\}$  on  $[0, 1]^2$  with uniform marginals, we denote

$$(3.3) \quad \sigma_J^2(\{G_l^*\}) = \lim_{n \rightarrow \infty} \left\{ \int_0^1 f^2(u) du + 2 \sum_{l=1}^n \int_0^1 \int_0^1 f(u)f(v) dG_l^*(u, v) \right\}$$

if the limit exists, where

$$(3.4) \quad f(u) = \int_0^1 (1_{[u \leq v]} - v) dJ(v) + J(u) - \int_0^1 J(v) dv$$

and  $J \in \mathcal{H}_\delta$  for some  $\delta > 0$ .

**THEOREM 3.1.** Suppose the sequence  $\{X_{ni}\}$  is  $\varphi$ -mixing with rate (3.1) or strong mixing with rate (3.2), the function  $g$  which defines the regression constants in (1.1) belongs to  $\mathbf{C}_{1,b}^*$  the space of functions which admit a derivative of bounded variation, and suppose that for any  $l > 1$ , there exists a continuous d.f.  $G_l^*$  on  $[0, 1]^2$  with uniform marginals such that

$$(3.5) \quad \lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} \left| F_{n,i,j}(F_n^{-1}(u), F_n^{-1}(v)) - G_{j-i}^*(u, v) \right| = 0$$

for all  $(u, v) \in [0, 1]^2$ .

Then, for every  $J \in \mathcal{G}_\delta$  with  $2 > \delta \geq 0$  if we have (3.1), and  $\delta > 0$  if we have (3.2)

$$(3.6) \quad \lim_{n \rightarrow \infty} D_2\left(\mathcal{L}(\mathcal{S}_n(J)), \mathcal{N}(0, \tilde{\sigma}_J^2(\{G_l^*\}))\right) = 0$$

where

$$(3.7) \quad \begin{aligned} \tilde{\sigma}_J^2(\{G_l^*\}) &= \sigma_J^2(\{G_l^*\}) \\ &\times \left( \int_0^1 \int_0^1 (u \wedge v) g'(u) g'(v) du dv - 2 \int_0^1 u g'(u) du + g^2(1) \right) \end{aligned}$$

and  $\tilde{\sigma}_J^2(\{G_l^*\}) < \infty$ .

*Remark 3.1.* Let the sequence of distribution functions  $\{F_{n,i,j}\}$  satisfy the following conditions:

(i) There exists a sequence of d.f.'s  $F_l^*$  and  $\mathbf{R}^2$  such that

$$\lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} |F_{n,i,j}(x_1, x_2) - F_{j-i}^*(x_1, x_2)| = 0 \quad \text{for all } (x_1, x_2) \in \mathbf{R}^2,$$

(ii)  $F_{ni} = F_n$  for all  $1 \leq i \leq n$ ,  $n \geq 1$ ,

then the condition (3.5) is satisfied whenever the sequence  $\{X_{ni}\}$  is strong mixing.

*Proof.* To prove Theorem 3.1, we first need a few lemmas.

For any  $n$  ( $n \geq 1$ ), and  $i$  ( $1 \leq i \leq n$ ) and any  $J \in \mathbf{C}_{2,b}$ , let

$$(3.8) \quad \begin{aligned} A_{ni}(J) &= \int_0^1 \left( \mathbf{1}_{[F_n(X_{ni}) \leq t]} - G_{ni}(t) \right) J'(t) dG_{ni}(t) \\ &+ J(F_n(X_{ni})) - \int_0^1 J(t) dG_{ni}(t). \end{aligned}$$

It is obvious that  $\mathbf{E}(A_{ni}) = 0$ .

Now consider for any  $J \in \mathbf{C}_{2,b}$ , the process  $L_n(J)(s)$  defined on  $\mathbf{C}_1$  the space of continuous functions on  $[0, 1]$ , by

$$(3.9) \quad L_n(J)(s) = n^{-1/2} \left( \sum_{i=1}^{[ns]} A_{ni} + (ns - [ns]) A_{n,[ns]+1} \right),$$

where  $[ns]$  denotes the integer part of the real number  $ns$ .

**LEMMA 3.1.** Suppose that  $\{X_{ni}\}$  satisfies the conditions of Theorem 3.1 and  $J$  belongs to  $\mathbf{C}_{2,b}$ , then the process  $L_n(J)(s)$  converges weakly in uniform topology to a Gaussian process  $L_0(J)(s)$  with trajectories a.s. in  $\mathbf{C}_1$  with mean 0 and variance  $\sigma_J^2(\{G_l^*\})$  where  $\sigma_J^2(\{G_l^*\})$  is defined in (3.3), and  $\sigma_J^2(\{G_l^*\}) < \infty$ .

*Proof.* The process  $L_n(J)$  defines a probability measure  $\mathbf{P}_n$  on  $\mathbf{C}_1$ . From Theorem 8.1 of [2] we have to prove that (i) the finite-dimensional distribution of  $\mathbf{P}_n$  converges in law to a normal distribution and (ii)  $\mathbf{P}_n$  is tight.

First we prove (i), which is equivalent to proving that  $\sum_{l=1}^p \lambda_l L_n(J)(s_l)$  converges in law to a normal distribution for any  $p \in \mathbf{N}^*$ , any  $l \in [0, 1]$  and any  $\lambda_l \in \mathbf{R}$  ( $1 \leq l \leq p$ ). Without loss of generality, we can take  $p = 2$  and suppose that  $s_1 < s_2$ .

We have

$$\begin{aligned}
 \sum_{l=1}^2 \lambda_l L_n(J)(s_l) &= n^{-1/2} \left[ \sum_{i=1}^{[ns_1]} (\lambda_1 + \lambda_2) A_{ni}(J) + \sum_{i=[ns_1]+1}^{[ns_2]} \lambda_2 A_{ni}(J) \right. \\
 &\quad \left. + \lambda_1 (ns_1 - [ns_1]) A_{n,[ns_1]+1}(J) \right. \\
 &\quad \left. + \lambda_2 (ns_2 - [ns_2]) A_{n,[ns_2]+1}(J) \right].
 \end{aligned}
 \tag{3.10}$$

We define the sequence of r.v.'s  $\{B_{ni}(J)\}$  by

$$B_{ni}(J) = \begin{cases} (\lambda_1 + \lambda_2) A_{ni}(J) & \text{if } i \leq [ns_1], \\ \lambda_2 A_{ni}(J) & \text{if } [ns_1] < i \leq [ns_2], \\ 0 & \text{if } i > [ns_2]. \end{cases}
 \tag{3.11}$$

As  $J$  and  $J'$  are bounded, we deduce

$$\sum_{l=1}^2 \lambda_l L_n(J)(s_l) = n^{-1/2} \sum_{i=1}^n B_{ni}(J) + O(n^{-1/2}).
 \tag{3.12}$$

From [11, Corollary 1] we have to verify that

$$\begin{aligned}
 \mathbf{E} \frac{1}{n} \left( \sum_{i=1}^n B_{ni}(J) \right)^2 &\longrightarrow \left\{ (\lambda_1 + \lambda_2)^2 s_1 + \lambda_2^2 (s_2 - s_1) \right\} \sigma_J^2(\{G_l^*\}) \\
 \text{as } n &\rightarrow \infty.
 \end{aligned}
 \tag{3.13}$$

We have

$$\begin{aligned}
 \mathbf{E} \frac{1}{n} \left( \sum_{i=1}^n B_{ni}(J) \right)^2 &= \frac{1}{n} \left[ (\lambda_1 + \lambda_2)^2 \sum_{i=1}^{[ns_1]} \sum_{j=1}^{[ns_1]} \mathbf{E} (A_{ni}(J) A_{nj}(J)) \right. \\
 &\quad \left. + (\lambda_1 + \lambda_2) \lambda_2 \sum_{i=1}^{[ns_1]} \sum_{j=[ns_1]+1}^{[ns_2]} \mathbf{E} (A_{ni}(J) A_{nj}(J)) \right. \\
 &\quad \left. + \lambda_2^2 \sum_{i=[ns_1]+1}^{[ns_2]} \sum_{j=[ns_1]+1}^{[ns_2]} \mathbf{E} (A_{ni}(J) A_{nj}(J)) \right].
 \end{aligned}
 \tag{3.14}$$

Suppose the sequence  $\{X_{ni}\}$  is  $\varphi$ -mixing, then from the boundedness of  $J$  and  $J'$  and from the well-known inequality on the moment of  $\varphi$ -mixing r.v.'s (see [5, Prop. 2.2]), we obtain

$$\begin{aligned}
 &\frac{1}{n} \left| (\lambda_1 + \lambda_2) \lambda_2 \sum_{i=1}^{[ns_1]} \sum_{j=[ns_1]+1}^{[ns_2]} \mathbf{E} (A_{ni}(J) A_{nj}(J)) \right| \\
 &\leq \frac{M}{n} \sum_{i=1}^{[ns_1]} \sum_{j=[ns_1]+1}^{[ns_2]} \varphi^{1/p}(j-i)
 \end{aligned}
 \tag{3.15}$$

where  $M$  is some positive constant and  $p = (4 + 2\delta)/(2 + 3\delta)$ . From (3.1), the last expression goes to zero as  $n \rightarrow \infty$ .

If the sequence  $\{X_{ni}\}$  is strong mixing, the left-hand side of (3.15) is majorized by

$$\frac{M'}{n} \sum_{i=1}^{[ns_1]} \sum_{j=[ns_1]+1}^{[ns_2]} \alpha^{\delta/(2+\delta)}(j-i),$$

where  $M' > 0$  is some constant and from (3.2), this converges to 0 as  $n \rightarrow \infty$ .

It remains to prove that

$$(3.16) \quad \begin{aligned} & \frac{1}{n} \sum_{i=1}^{[ns_1]} \sum_{j=1}^{[ns_1]} \mathbf{E}(A_{ni}(J) A_{nj}(J)) \longrightarrow s_1 \sigma_J^2(\{G_l^*\}) \\ & \text{as } n \rightarrow \infty, \\ & \frac{1}{n} \sum_{i=[ns_1]+1}^{[ns_2]} \sum_{j=[ns_1]+1}^{[ns_2]} \mathbf{E}(A_{ni}(J) A_{nj}(J)) \longrightarrow (s_2 - s_1) \sigma_J^2(\{G_l^*\}) \\ & \text{as } n \rightarrow \infty. \end{aligned}$$

We first prove the convergence of (3.16) when  $J$  and  $J'$  are replaced by indicator functions. Suppose

$$J(t) = \mathbf{1}_{[a \leq t \leq b]} \quad \text{and} \quad J'(t) = \mathbf{1}_{[a' \leq t \leq b']}.$$

Then we can write

$$\begin{aligned} A_{ni}(J) &= G_{ni}(b') - G_{ni}\left(\{a' \vee F_n(X_{ni})\} \wedge b'\right) \\ &\quad - \int_a^b t dG_{ni}(t) + \mathbf{1}_{[a \leq F_n(X_{ni}) \leq b]} - (G_{ni}(b) - G_{ni}(a)) = D_{ni}(X_{ni}^*) \end{aligned}$$

where  $X_{ni}^* = F_n(X_{ni})$ . Let  $G_{n,i,j}$  be the d.f. of  $(F_n(X_{ni}), F_n(X_{nj}))$ . Then, we have

$$\mathbf{E}(A_{ni}(J) A_{nj}(J)) = \int_0^1 \int_0^1 D_{ni}(u) D_{nj}(v) dG_{n,i,j}(u, v).$$

From condition (3.5) we easily deduce that

$$(3.17) \quad \begin{aligned} & \lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} \left| \mathbf{E}(A_{ni}(J) A_{nj}(J)) \right. \\ & \quad \left. - \int_0^1 \int_0^1 D(u) D(v) dG_{j-i}^*(u, v) \right| = 0 \end{aligned}$$

where

$$D(u) = \mathbf{1}_{[a \leq u \leq b]} - (b - a) + b' - \{(a' \vee u) \wedge b'\} - \frac{1}{2}(b' - a')^2.$$

We obtain the same result if  $J$  and  $J'$  are replaced by step functions.

As  $J$  and  $J'$  are continuous and bounded, we can uniformly approach them by step functions and we deduce that

$$(3.18) \quad \lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} \left| \mathbf{E}(A_{ni}(J) A_{nj}(J)) - \int_0^1 \int_0^1 f(u) f(v) dG_{j-i}^*(u, v) \right| = 0$$



where  $f(u)$  is defined in (3.4).

Now denote  $\rho(0) = \int_0^1 f^2(u)du$  and  $\rho(i) = 2 \int_0^1 \int_0^1 f(u)f(v) dG_i^*(u, v)$ ,  $i \geq 1$ . Then,

$$\begin{aligned} & \left| n^{-1} \sum_{i=1}^{[ns_1]} \sum_{j=1}^{[ns_1]} \mathbf{E}(A_{ni}(J) A_{nj}(J)) - [ns_1] n^{-1} \sum_{i=0}^{+\infty} \rho(i) \right| \\ & \leq \left| n^{-1} [ns_1] ([ns_1])^{-1} \sum_{i=0}^{[ns_1]-1} \sum_{j=1}^{[ns_1]-i} \mathbf{E}(A_{nj}(J) A_{n,j+i}(J)) \right. \\ & \quad \left. - [ns_1] (n[ns_1])^{-1} \sum_{i=0}^{[ns_1]} ([ns_1] - i) \rho(i) \right| + [ns_1] n^{-1} \sum_{i=[ns_1]+1}^{\infty} |\rho(i)| \\ & \quad + [ns_1] n^{-1} \sum_{i=0}^{[ns_1]} \sum_{k=i}^{\infty} |\rho(k)| = |A_n| + B_n + C_n. \end{aligned}$$

From (3.18) we deduce that  $|A_n| \rightarrow 0$  as  $n \rightarrow \infty$  and from the well-known inequalities on the moment of mixing r.v.'s (see [5, Props. 2.2 and 2.8]) and (3.1) or (3.2) we deduce that  $B_n \rightarrow 0$  and  $C_n \rightarrow 0$  and  $n \rightarrow \infty$ .

It is also immediate that

$$\left| [ns_1] n^{-1} \sum_{i=0}^{+\infty} \rho(i) - s_1 \sum_{i=0}^{+\infty} \rho(i) \right| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

We conclude that  $n^{-1} \sum_{i=1}^{[ns_1]} \sum_{j=1}^{[ns_1]} \mathbf{E}(A_{ni}(J) A_{nj}(J))$  converges to  $s_1 (\sum_{i=0}^{+\infty} \rho(i))$  as  $n \rightarrow \infty$  where  $\sum_{i=0}^{+\infty} \rho(i)$  is equal to  $\sigma_J^2(\{G_l^*\})$ . Similarly

$$\frac{1}{n} \sum_{i=[ns_1]+1}^{[ns_2]} \sum_{j=[ns_1]+1}^{[ns_2]} \mathbf{E}(A_{ni}(J) A_{nj}(J)) \rightarrow (s_2 - s_1) \sigma_J^2(\{G_l^*\})$$

as  $n \rightarrow \infty$ , so (3.16) is proved.

From (3.14)–(3.16), we deduce (3.13) and we conclude that  $\mathbf{E}(\sum_{l=1}^2 \lambda_l L_n(J)(s_l))^2$  converges to  $\{(\lambda_1 + \lambda_2)^2 s_1 + \lambda_2^2 (s_2 - s_1)\} \sigma_J^2(\{G_l^*\})$  which implies that  $\sum_{l=1}^2 \lambda_l L_n(J)(s_l)$  converges in law to the normal distribution with mean 0 and variance  $\{(\lambda_1 + \lambda_2)^2 s_1 + \lambda_2^2 (s_2 - s_1)\} \sigma_J^2(\{G_l^*\})$  and (i) is proved.

We now prove (ii).

From [2, Thm. 8.2] we have to verify that  $\forall \varepsilon > 0 \exists \eta > 0$  ( $0 < \eta < 1$ ) and an integer  $N_0$  such that  $\forall n \geq N_0$

$$(3.19) \quad \mathbf{P} \left[ \sup_{|s-s'| < \eta} |L_n(J)(s) - L_n(J)(s')| \geq \varepsilon \right] \leq \varepsilon.$$

If  $ns$  and  $ns'$  are integers, by using Theorem 11 from [5] for  $q = 4$  for the strong mixing case and Lemma 5.1 in Harel (1988) for  $q = 2$  for the  $\varphi$ -mixing case, we obtain for  $s \geq s'$

$$(3.20) \quad \mathbf{E}(L_n(J)(s) - L_n(J)(s'))^4 \leq ((s - s')^2 + n^{-1}(s - s')) MC(\beta)$$

where

$$(3.21) \quad C(\beta) = \sum_{m=1}^{+\infty} m^{-1} \varphi^{1/4}(m)$$

if the sequence  $\{X_{ni}\}$  is  $\varphi$ -mixing, and

$$(3.22) \quad C(\beta) = \sum_{m=1}^{+\infty} m^2 \alpha^{\delta/(2+\delta)}(m)$$

if the sequence  $\{X_{ni}\}$  is strong mixing and  $M$  is some positive constant.

If  $s > s'$  and  $ns$  and  $ns'$  are integers, we have  $s - s' \geq n^{-1}$  and

$$\mathbf{E} (L_n(J)(s) - L_n(J)(s'))^4 \leq 2M(s - s')^2 C(\beta).$$

From [1, Lemma 2] we obtain that for any  $\varepsilon > 0$  there exist  $\eta > 0$ , and an integer  $N_0$  sufficiently large such that  $\forall n \geq N_0$ ,

$$(3.23) \quad \begin{aligned} & \mathbf{P} \left[ \sup_{|[ns]/n - [ns']/n| < 2\eta} |L_n(J)([ns]/n) - L_n(J)([ns']/n)| > \varepsilon/2 \right] \\ & \leq 2MC(\beta)K\eta^2\varepsilon^{-4} \end{aligned}$$

where  $K$  is some positive constant.

From the definition of  $L_n(J)(s)$  in (3.9), we obtain

$$(3.24) \quad \begin{aligned} & \sup_{|s-s'| < \eta} |L_n(J)(s) - L_n(J)(s')| \\ & \leq 2 \max_{|[ns]/n - [ns']/n| < 2\eta} |L_n(J)([ns]/n) - L_n(J)([ns']/n)|. \end{aligned}$$

By using (3.23) and (3.24), we deduce

$$(3.25) \quad \mathbf{P} \left[ \sup_{|s-s'| < \eta} |L_n(J)(s) - L_n(J)(s')| \geq \varepsilon \right] \leq MC(\beta)K'\eta^2\varepsilon^{-4}$$

where  $K'$  is some positive constant and (3.19) is proved. The fact that  $\sigma_J^2(\{G_l^*\}) < \infty$  is a simple consequence of  $J \in \mathbf{C}_{2,b}$  and (3.1) or (3.2).

Now we consider for any  $J \in \mathbf{C}_{2,b}$  the r.v.  $V_n(J)$  defined by

$$(3.26) \quad V_n(J) = n^{-1/2} \sum_{i=1}^n c_{ni} A_{ni}.$$

**LEMMA 3.2** *Suppose that  $\{X_{ni}\}$  satisfies the conditions of Theorem 3.1,  $J$  belongs to  $\mathbf{C}_{2,b}$  and  $g$  admits a derivative  $g'$ . Then  $V_n(J)$  converges in law to the normal distribution with mean 0 and variance  $\tilde{\sigma}_J^2(\{G_l^*\})$  where  $\tilde{\sigma}_J^2(\{G_l^*\})$  is defined in (3.7) and  $\tilde{\sigma}_J^2(\{G_l^*\}) < \infty$ .*

*Proof.* For any  $n$  define a measure  $\lambda_n$  on  $[0, 1]$  by setting

$$\lambda_n(\{i/n\}) = c_{ni} - c_{n,i+1}, \quad 1 \leq i \leq n-1, \quad \text{and} \quad \lambda_n(\{1\}) = c_{nn}.$$

By definition, we have

$$V_n(J) = \int_0^1 L_n(J)(u) \lambda_n(du).$$

We now prove that

$$(3.27) \quad \int_0^1 L_n(J)(u) \lambda_n(du) \quad \text{converges in law to} \\ - \int_0^1 L_0(J)(u) g'(u) du + L_0(J)(1)g(1) \quad \text{as } n \rightarrow \infty.$$

Let  $h_n: \mathbf{C}_1 \rightarrow \mathbf{R}$ ,  $n \geq 1$ , be defined as  $h_n(f) = \int_0^1 f(u) \lambda_n(du)$  and  $h_0: \mathbf{C}_1 \rightarrow \mathbf{R}$  be defined as  $h_0(f) = - \int_0^1 f g'(u) du + f(1)g(1)$ . Let  $\{f_n, n \geq 1\}$  be a sequence of functions in  $\mathbf{C}_1$  and suppose that  $f_n \rightarrow f_0$  in uniform topology as  $n \rightarrow \infty$  where  $f_0 \in \mathbf{C}_1$ . We show that  $h_n(f_n) \rightarrow h_0(f_0)$  as  $n \rightarrow \infty$ .

$$\begin{aligned} & \left| \int_0^1 f_n(u) \lambda_n(du) - \left( - \int_0^1 f_0(u) g'(u) du + f_0(1)g(1) \right) \right| \\ & \leq \left| \int_0^1 (f_n(u) - f_0(u)) \lambda_n(du) \right| \\ & \quad + \left| \int_0^1 f_0(u) \lambda_n(du) + \int_0^1 f_0(u) g'(u) du - f_0(1)g(1) \right| \\ & \leq \sup_{u \in [0,1]} |f_n(u) - f_0(u)| \left| \int_0^1 \lambda_n(du) \right| \\ & \quad + \left| \sum_{i=1}^{n-1} f_0(i/n) \left( g(i/n) - g((i+1)/n) \right) + f_0(1)g(1) \right. \\ & \quad \left. + \int_0^1 f_0(u) g'(u) du - f_0(1)g(1) \right| \rightarrow 0 \quad \text{as } n \rightarrow \infty \end{aligned}$$

from the hypothesis  $f_n \rightarrow f_0$  in uniform topology,  $g'$  is an integrable function and  $\sup_{n \in \mathbf{N}^*} \Lambda_n < \infty$ .

Consequently,  $h_n(f_n) \rightarrow h_0(f_0)$  as  $n \rightarrow \infty$  and by [2, Thm. 5.5] (3.27) follows.

It remains to show that

$$\mathbf{E} \left( - \int_0^1 L_0(J)(u) g'(u) du + L_0(J)(1)g(1) \right)^2 = \tilde{\sigma}_J(\{G_l^*\}) < \infty.$$

We have

$$\begin{aligned} & \mathbf{E} \left( - \int_0^1 L_0(J)(u) g'(u) du + L_0(J)(1)g(1) \right)^2 \\ & = \int_0^1 \int_0^1 \mathbf{E} [L_0(J)(u) L_0(J)(v)] g'(u) g'(v) du dv \\ & \quad - 2 \int_0^1 \mathbf{E} [L_0(J)(1) L_0(J)(u)] g'(u) du + \mathbf{E} [L_0(J)(1) L_0(J)(1)] g^2(1). \end{aligned}$$

As  $\mathbf{E} [L_0(J)(u) L_0(J)(v)] = (u \wedge v) \sigma_J^2(\{G_l^*\}) < \infty$ , the property follows and Lemma 3.2 is proved.

*Proof of the theorem.* We first prove that the theorem is true for  $J \in \mathbf{C}_{2,b}$ . We have the following decomposition:

$$\mathcal{S}_n(J) = V_n(J) + A_n(J) + B_n(J) + C_n(J),$$

where

$$\begin{aligned} A_n(J) &= n^{1/2} \int_0^1 J'(I_n(t)) (\hat{I}_n(t) - I_n(t)) d(\hat{G}_n - G_n)(t), \\ B_n(J) &= -(n+1)^{-1} n^{1/2} \int_0^1 J'(I_n(t)) \hat{I}_n(t) d\hat{G}_n(t), \\ C_n(J) &= 2^{-1} n^{1/2} \int_0^1 J''(\theta_n(I_n(t))) \left( n(n+1)^{-1} (\hat{I}_n(t) - I_n(t)) \right)^2 d\hat{G}_n(t), \end{aligned}$$

where  $\theta_n(I_n(t)) \in [I_n(t) \wedge \hat{I}_n(t), I_n(t) \vee \hat{I}_n(t)]$ .

Suppose  $J \in \mathbf{C}_{2,b}$ . Then the weak convergence of  $V_n(J)$  is established in Lemma 3.2. The random variables  $A(J)$ ,  $B(J)$ ,  $C(J)$  converge to zero in probability and in  $\mathbf{L}_2$ , since

$$(3.28) \quad \mathbf{E} (A_n^2(J)) \leq K n^{-1} \sup_{t \in [0,1]} |J'(t)|^2 \sup_{n \in \mathbf{N}^*} \Lambda_n^2 C(\beta),$$

$$(3.29) \quad |B_n(J)| \leq K n^{-1/2} \sup_{t \in [0,1]} |J'(t)| \sup_{n \in \mathbf{N}^*} \Lambda_n,$$

$$(3.30) \quad \mathbf{E} (C_n^2(J)) \leq K n^{-1} \sup_{t \in [0,1]} |J'(t)|^2 \sup_{n \in \mathbf{N}^*} \Lambda_n^2 C(\beta)$$

where  $C(\beta) = \sum_{m=1}^{+\infty} m(\varphi(m))^{(2+3\delta)/(4+2\delta)}$  if we have (3.1), and  $C(\beta) = m^2 \times \sum_{m=1}^{+\infty} (\alpha(m))^{\delta/(2+\delta)}$  if we have (3.2), and  $K$  is some positive constant.

We only prove the inequality (3.28) for the  $\varphi$ -mixing case, because the method is similar to the proof of the three inequalities in [4, p. 66]. For any  $(i, j, l, q) \in \mathbf{N}^4$ , we put

$$\begin{aligned} \beta(i, j, l, q) &= \left( \int_0^1 J'(I_n(t)) c_{ni} (\mathbf{1}_{[X_{ni} \leq t]} - F_{ni}(t)) d(\mathbf{1}_{[X_{nj} \leq t]} - F_{nj}(t)) \right) \\ &\quad \times \left( \int_0^1 J'(I_n(u)) c_{nl} (\mathbf{1}_{[X_{nl} \leq u]} - F_{nl}(u)) d(\mathbf{1}_{[X_{nq} \leq u]} - F_{nq}(u)) \right). \end{aligned}$$

Suppose  $i \leq j \leq l \leq q$  and let  $p = (2 + 3\delta)/(4 + 2\delta)$ , then from the condition of  $\varphi$ -mixing, we have the following three inequalities:

$$\beta(i, j, l, q) \leq \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \varphi(j-i),$$

$$\beta(i, j, l, q) \leq \left( 2 \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \right) \left( \varphi(l-j) + 4\varphi^{1/p}(j-i)\varphi^{1/p}(q-l) \right),$$

$$\beta(i, j, l, q) \leq \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \varphi(q-l).$$

If  $i, j, l, q$  are differently ordered, we obtain similar inequalities.

From this, we deduce

$$\mathbf{E} (A_n^2(J)) \leq 4! n^{-3} \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \sum_{i,j,l,q} \beta(i, j, l, q).$$

Put  $j' = j - i$ ,  $l' = l - j$  and  $q' = q - l$ . We have

$$\begin{aligned} \mathbf{E}(A_n^2(J)) &\leq 4! n^{-3} \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \\ &\quad \times \sum_{i=1}^n \left( \sum_{\substack{0 \leq l' \leq j' \\ 0 \leq q' \leq j'}} 2\varphi(j') + \sum_{\substack{0 \leq j' \leq q' \\ 0 \leq l' \leq q'}} 2\varphi(q') + \sum_{\substack{0 \leq j' \leq l' \\ 0 \leq q' \leq l'}} 2\varphi(l') + 4\varphi^{1/p}(j')\varphi^{1/p}(q') \right) \end{aligned}$$

and after some computations, we obtain

$$\begin{aligned} \mathbf{E}(A_n^2(J)) &\leq 288n^{-3} \sup_{t \in [0,1]} |J'(t)|^2 \Lambda_n^2 \left( \sum_{m \geq 1} \varphi^{1/p}(m) \right) \left( \sum_{m \geq 1} m\varphi^{1/p}(m)/m \right) (n + n^2) \\ &\leq Kn^{-1} \sup_{t \in [0,1]} |J'(t)|^2 \sup_{n \in \mathbf{N}^*} \Lambda_n^2 C(\beta) \end{aligned}$$

where  $K$  is some positive constant and  $C(\beta) = \sum_{m \geq 1} m(\varphi(m))^{(2+3\delta)/(4+2\delta)}$ . The inequality (3.28) is proved for the  $\varphi$ -mixing case. Hence the theorem is true for  $J \in \mathbf{C}_{2,b}$ .

Following Proposition 2.3 it remains to prove that the operator  $\sigma: \mathcal{H}_\delta \rightarrow \mathbf{R}$  defined by  $\sigma(J) = \tilde{\sigma}_J(\{G_l^*\})$  satisfies the Lipschitz condition for the  $\|\cdot\|_\delta$  norm and the condition (2.5) is satisfied. The first property follows easily from the definition of  $\tilde{\sigma}_J(\{G_l^*\})$  in (3.7) and the definition of  $\sigma_J(\{G_l^*\})$  in (3.3) and (3.4).

We now prove (2.5) if we have (3.1). For  $p = (4+2\delta)/(2+3\delta)$  and  $q = (1-p^{-1})^{-1}$  we have

$$\begin{aligned} n \mathbf{E}(\hat{G}_n(t) - G_n(t))^2 &= n \mathbf{E} \left[ n^{-1} \sum_{i=1}^n c_{ni} (\mathbf{1}_{[F_n(X_{ni}) \leq t]} - G_{ni}(t)) \right]^2 \\ &= n^{-1} \sum_{1 \leq i, j \leq n} \mathbf{E} \left[ c_{ni} c_{nj} (\mathbf{1}_{[F_n(X_{ni}) \leq t]} - G_{ni}(t)) (\mathbf{1}_{[F_n(X_{nj}) \leq t]} - G_{nj}(t)) \right] \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} \sum_{j=1}^{n-i} \mathbf{E} \left| (\mathbf{1}_{[F_n(X_{ni}) \leq t]} - G_{ni}(t)) (\mathbf{1}_{[F_n(X_{n,j+i}) \leq t]} - G_{n,j+i}(t)) \right| \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} \varphi^{1/p}(i) \sum_{j=1}^{n-i} \left( G_{n,j}(t) (1 - G_{n,j}(t)) \right)^{1/p} \\ &\quad \times \left( G_{n,j+i}(t) (1 - G_{n,j+i}(t)) \right)^{1/q} \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=1}^{n-1} \varphi^{1/p}(i) \left( \sum_{j=1}^n G_{n,j}(t) (1 - G_{n,j}(t)) \right)^{1/p} \\ &\quad \times \left( \sum_{j=1}^n G_{n,j}(t) (1 - G_{n,j}(t)) \right)^{1/q} \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=1}^{n-1} \varphi^{1/p}(i) \left( \sum_{j=1}^n G_{n,j}(t) (1 - G_{n,j}(t)) \right) \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=1}^{n-1} \varphi^{1/p}(i) n \left( \frac{1}{n} \sum_{j=1}^n G_{n,j}(t) \right) \left( 1 - n^{-1} \sum_{j=1}^n G_{n,j}(t) \right) \end{aligned}$$

$$= 2\Lambda_n^2 \sum_{i=1}^{n-1} \varphi^{1/p}(i)t(1-t) \leq 2 \left( \sum_{i=1}^{+\infty} \varphi^{1/p}(i) \right) \Lambda_n^2 t(1-t),$$

which implies (2.5) if we have (3.1).

Finally, we prove (2.5) if we have (3.2). We have in analogy with preceding arguments

$$\begin{aligned} n \mathbf{E} \left( \widehat{G}(t) - G_n(t) \right)^2 &\leq 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} (\alpha(i))^{\delta/(2+\delta)} \sum_{j=1}^{n-i} \left( G_{nj}(t) (1 - G_{nj}(t)) \right)^{1/(2+\delta)} \\ &\quad \left( G_{n,j+i}(t) (1 - G_{n,j+i}(t)) \right)^{1/(2+\delta)} \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} (\alpha(i))^{\delta/(2+\delta)} \left( \sum_{j=1}^n G_{nj}(1 - G_{nj}(t)) \right)^{1/(2+\delta)} \\ &\quad \times \left( \sum_{j=1}^n G_{n,j}(t) (1 - G_{n,j}(t)) \right)^{1/(2+\delta)} \\ &= 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} (\alpha(i))^{2/(2+\delta)} \left( \sum_{j=1}^n G_{nj}(t) (1 - G_{nj}(t)) \right)^{2/(2+\delta)} \\ &\leq 2n^{-1} \Lambda_n^2 \sum_{i=0}^{n-1} (\alpha(i))^{\delta/(2+\delta)} n \left( n^{-1} \sum_{j=1}^n G_{nj}(t) \left( 1 - n^{-1} \sum_{j=1}^n G_{nj}(t) \right) \right)^{2/(2+\delta)} \\ &= 2\Lambda_n^2 \sum_{i=0}^{n-1} (\alpha(i))^{\delta/(2+\delta)} (t(1-t))^{2/(2+\delta)} \\ &\leq 2\Lambda_n^2 \sum_{i=0}^{+\infty} (\alpha(i))^{2/(2+\delta)} (t(1-t))^{1-2\eta} \quad \text{for } \eta = \delta(4+2\delta)^{-1}. \end{aligned}$$

Thus (2.5) is proved if we have (3.2).

**COROLLARY 3.1.** *If among conditions in Theorem 3.1 the function  $g$  is replaced by a function for which there exists a decomposition  $g = g_c + g_d$  where  $g_c \in \mathbf{C}_{1,b}^*$  and  $g_d$  is a step function with  $p$  jumps, say at  $a_1, \dots, a_p$  such that  $a_i \in (0, 1)$  ( $1 \leq i \leq p$ ), then the conclusion of Theorem 3.1 remains true but  $\tilde{\sigma}_J(\{G_l^*\})$  defined in (3.7) is replaced by  $\hat{\sigma}_J^2(\{G_l^*\})$  where*

$$\begin{aligned} \hat{\sigma}_J^2(\{G_l^*\}) &= \sigma_J^2(\{G_l^*\}) \left( \int_0^1 \int_0^1 (u \wedge v) g'_c(u) g'_c(v) du dv \right. \\ &\quad - 2 \sum_{i=1}^p (g_d(a_i-) - g_d(a_i+)) \int_0^1 (u \wedge a_i) g'_c(u) du \\ &\quad - 2 \int_0^1 u g'_c(u) du + \sum_{1 < i, j \leq p} (a_i \wedge a_j) (g_d(a_i-) \\ &\quad \left. - g_d(a_i+)) (g_d(a_j-) - g_d(a_j+)) + 2 \sum_{i=1}^p a_i (g_d(a_i-) - g_d(a_i+)) + g^2(1) \right), \end{aligned} \tag{3.31}$$

where  $\sigma_J^2(\{G_l^*\})$  is defined in (3.3).

**4. Convergence of the two-sample linear rank statistic.** Let  $\{Y_{n1i}\}$ ,  $1 \leq i \leq n_1$ , and  $\{Z_{n2j}\}$ ,  $1 \leq j \leq n_2$ , be two independent sequences of weakly dependent random variables with continuous d.f.'s  $F_{n1i}^{(1)}(x)$  and  $F_{n2j}^{(2)}(x)$ , respectively,  $x \in \mathbf{R}$ . Given  $n = n_1 + n_2$  we set  $X_{ni} = Y_{n1i}$  if  $i \leq n_1$  and  $X_{ni} = Y_{n2i-n_1}$  if  $i > n_1$ . Denote by  $\hat{F}_{n1}^{(1)}(x) = n_1^{-1} \sum_{i=1}^{n_1} c_{n1i} \mathbf{1}_{[X_{ni} \leq x]}$  the empirical process based on the first sequence of r.v.'s  $\{Y_{n1i}\}$  and weighted by the regression constants  $c_{n1i}$ . We put  $F_{n1}^{(1)} = \mathbf{E}(\hat{F}_{n1}^{(1)})$ .

Then  $\mathcal{S}_n^*(J)$  defined by

$$(4.1) \quad \mathcal{S}_n^*(J) = n^{1/2} \left( \int_{-\infty}^{+\infty} J \left( \frac{n}{n+1} \hat{F}_n(x) \right) d\hat{F}_{n1}^{(1)}(x) - \int_{-\infty}^{+\infty} J(F_n(x)) dF_{n1}^{(1)}(x) \right)$$

is the two-sample linear rank statistic. We suppose that the regression constants  $c_{n1i}$  ( $1 \leq i \leq n_1$ ) are defined by a function  $h$  on  $[0, 1]$  as

$$c_{n1i} = h(i/n_1), \quad 1 \leq i \leq n_1, \quad n_1 \geq 1.$$

We assume that  $n_1 n^{-1} \rightarrow \lambda_0 \in (0, 1)$ .

We have  $F_n = n^{-1} (\sum_{i=1}^{n_1} F_{n1i}^{(1)} + \sum_{j=1}^{n_2} F_{n2j}^{(2)})$ .

Let  $F_{n1,i,l}^{(1)}$  be the d.f. of  $(Y_{n1i}, Y_{n1l})$  and  $F_{n2,j,k}^{(2)}$  be the d.f. of  $(Z_{n2j}, Z_{n2k})$ .

**THEOREM 4.1.** *Suppose the sequences  $\{Y_{n1i}\}$  and  $\{Z_{n2j}\}$  are  $\varphi$ -mixing with rate (3.1) or strong mixing with rate (3.2), the function  $h$  satisfies  $h = h_c + h_d$  with  $h_c \in \mathbf{C}_{1,b}^*$  and  $h_d$  is a step function and if for each  $p > 1$ , there exist two continuous d.f.'s  $\hat{G}_p^{(1)}$  and  $\hat{G}_p^{(2)}$  on  $\mathbf{R}^2$  with marginals  $F^{(1)}$  and  $F^{(2)}$  such that*

$$(4.2) \quad \lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} \left| F_{n1,i,j}^{(l)}(F_n^{-1}(t_1), F_n^{-1}(t_2)) - \hat{G}_{j-i}^{(l)}(H^{-1}(t_1), H^{-1}(t_2)) \right| = 0$$

for all  $(t_1, t_2) \in [0, 1]^2$ ,  $l = 1, 2$ , where

$$(4.3) \quad H = \lambda_0 F^{(1)} + (1 - \lambda_0) F^{(2)}.$$

Then, for every  $J \in \mathcal{G}_\delta$  with  $2 > \delta \geq 0$  if we have (3.1) and  $\delta > 0$  if we have (3.2)

$$(4.4) \quad \lim D_2 \left( \mathcal{L}(\mathcal{S}_n^*(J)), \mathcal{N}(0, \tilde{\sigma}_J(\{G_p^{(l)}\})) \right) = 0$$

where

$$(4.5) \quad \tilde{\sigma}_J^2(\{\hat{G}_p^{(l)}\}) = \tilde{\sigma}_J^2(\{\hat{G}_p^{(l)}\}) L(h)$$

where

$$(4.6) \quad \begin{aligned} \tilde{\sigma}_J^2(\{G_p^{(l)}\}) = & \lambda_0^{-1} \left\{ \int_0^1 f_1^2(u) d(F^{(1)} \circ H^{-1})(u) \right. \\ & + 2 \sum_{p \geq 2} \int_0^1 \int_0^1 f_1(u) f_1(v) d(\hat{G}_p^{(1)}(H^{-1}(u), H^{-1}(v))) \Big\} \\ & + (1 - \lambda_0) \left\{ \int_0^1 f_2^2(u) d(F^{(2)} \circ H^{-1})(u) \right. \\ & + 2 \sum_{p \geq 2} \int_0^1 \int_0^1 f_2(u) f_2(v) d(\hat{G}_p^{(2)}(H^{-1}(u), H^{-1}(v))) \Big\}. \end{aligned}$$

Here

$$\begin{aligned} f_1(u) &= f_1^*(u) - \int_0^1 f_1^*(v) d(F^{(1)} \circ H^{-1}(v)), \\ f_2(u) &= f_2^*(u) - \int_0^1 f_2^*(v) d(F^{(2)} \circ H^{-1}(v)), \end{aligned}$$

with

$$\begin{aligned} f_1^*(u) &= J(u) + \lambda_0 \int_u^1 J'(v) d(F^{(1)} \circ H^{-1}(v)), \\ f_2^*(u) &= \int_u^1 J'(v) d(F^{(2)} \circ H^{-1}(v)), \end{aligned}$$

and

$$\begin{aligned} L(h) &= \int_0^1 \int_0^1 h'_c(u) h'_c(v) du dv - 2 \sum_{i=1}^p (h_d(a_i-) - h_d(a_i+)) \int_0^1 h_c(u) du \\ &\quad - 2 \int_0^1 h'_c(u) du + \sum_{1 \leq i, j \leq p} (h_d(a_i-) - h_d(a_i+)) (h_d(a_j-) - h_d(a_j+)) \\ (4.7) \quad &+ 2 \sum_{i=1}^p (h_d(a_i-) - h_d(a_i+)) + h_c^2(1), \end{aligned}$$

where  $a_i$ ,  $1 \leq i \leq p$ , are the discontinuous points of  $h_d$ .

*Remark.* Let the sequences  $\{F_{n_1, i, j}^{(1)}\}$  and  $\{F_{n_2, i, j}^{(2)}\}$  satisfy the following conditions:

(i) there exist two sequences of d.f.'s  $G_p^{(l)}$  on  $\mathbf{R}^2$ ,  $l = 1, 2$ , such that

$$\lim_{n \rightarrow \infty} \max_{1 \leq i < j \leq n} |F_{n_1, i, j}^{(l)}(x_1, x_2) - G_{j-1}^{(l)}(x_1, x_2)| = 0$$

for all  $(x_1, x_2) \in \mathbf{R}^2$ ,  $l = 1, 2$ ,

(ii)  $F_{n_1 i}^{(1)} = F_{n_1}^{(1)}$  for all  $1 \leq i \leq n_1$ ,

(iii)  $F_{n_2 j}^{(2)} = F_{n_2}^{(2)}$  for all  $1 \leq j \leq n_2$ .

Then condition (4.2) is satisfied when the sequences  $\{Y_{n_1 i}\}$  and  $\{Z_{n_2 j}\}$  are strong mixing.

*Proof of Theorem 4.2.* For any  $n_1 \geq 1$ , for any  $i$  ( $1 \leq i \leq n_1$ ) and for any  $J \in \mathbf{C}_{2, b}$ , let

$$\begin{aligned} B_{n_1 i}(J) &= n_1 n^{-1} \int_0^1 \left( \mathbf{1}_{[F_n(Y_{n_1 i}) \leq t]} - G_{n_1 i}^{(1)}(t) \right) J'(t) dG_{n_1 i}^{(1)}(t) \\ (4.8) \quad &+ J(F_n(Y_{n_1 i})) - \int_0^1 J(t) dG_{n_1 i}^{(1)}(t) \end{aligned}$$

where  $G_{n_1 i}^{(1)} = F_{n_1 i}^{(1)} \circ F_n^{-1}$ , and for any  $n_2 \geq 1$ , any  $j$  ( $1 \leq j \leq n_2$ ), any  $J \in \mathbf{C}_{2, b}$  and any  $u \in [0, 1]$ , let

$$\begin{aligned} C_{n_2 j}(J)(u) &= n_2^{-1} \sum_{l=1}^{[n_2 u]} \int_0^1 \left( \mathbf{1}_{[F_n(Z_{n_2 j}) \leq t]} - G_{n_2 j}^{(2)}(t) \right) J'(t) dG_{n_2 l}^{(2)}(t) \\ (4.9) \quad &+ n_2^{-1} (n_2 u - [n_2 u]) \int_0^1 \left( \mathbf{1}_{[F_n(Z_{n_2 j}) \leq t]} - G_{n_2 j}^{(2)}(t) \right) dG_{n_2, [n_2 u] + 1}^{(2)}(t), \end{aligned}$$



where  $G_{n_2j}^{(2)} = F_{n_2j}^{(2)} \circ F_n^{-1}$ .

Now consider for any  $J \in \mathbf{C}_{2,b}$  the processes  $W_{n_1}(J)(s)$  and  $W_{n_2}^*(J)(s, u)$  defined, respectively, on  $\mathbf{C}_1$  and  $\mathbf{C}_1^*$  (= the space of continuous functions on  $[0, 1]^2$ ) by

$$(4.10) \quad W_{n_1}(J)(s) = n_1^{1/2} \left( \sum_{i=1}^{[n_1 s]} B_{n_1 i}(J) + (n_1 s - [n_1 s]) B_{n_1, [n_1 s] + 1}(J) \right),$$

$$(4.11) \quad W_{n_2}^*(J)(s, u) = n_2^{1/2} \left( \sum_{i=1}^{[n_2 s]} C_{n_2 j}(J)(u) + (n_2 s - [n_2 s]) C_{n_2, [n_2 s] + 1}(J)(u) \right).$$

By similar techniques as in Lemma 3.1 one can prove that the process  $W_{n_1}(J)(s)$  converges weakly in uniform topology to a Gaussian process  $W_0(J)(s)$  with trajectories a.s. in  $\mathbf{C}_1$  with mean 0 and variance

$$(4.12) \quad s \left[ \int_0^1 \int_0^1 f_1^2(u) d(F^{(1)} \circ H^{-1})(u) + 2 \sum_{p \geq 1} \int_0^1 \int_0^1 f_1(u) f_1(v) d(G_p^{(1)}(H^{-1}(u), H^{-1}(v))) \right]$$

and  $W_{n_2}^*(J)(s, u)$  converges weakly in uniform topology to a Gaussian process  $W_0^*(J)(s, u)$  with trajectories a.s. in  $\mathbf{C}_1^*$  with mean 0 and variance

$$(4.13) \quad su \left[ \int_0^1 f_2^2(u) d(F^{(2)} \circ H^{-1})(u) + 2 \sum_{p \geq 1} \int_0^1 \int_0^1 f_2(u) f_2(v) d(G_p^{(2)}(H^{-1}(u), H^{-1}(v))) \right].$$

From this and following Lemma 3.2, it is easy to prove that  $V_n^*(J)$  converges in law to the normal distribution with mean 0 and variance  $\tilde{\sigma}_J(\{\hat{G}_p^{(l)}\})$  where  $\tilde{\sigma}_J(\{\hat{G}_p^{(l)}\})$  is defined in (4.5) and  $V_n^*(J)$  is a random variable defined by

$$(4.14) \quad V_n^*(J) = n^{1/2} \left( \int_{-\infty}^{+\infty} J(\hat{F}_n(x)) d(\hat{F}_n^{(1)}(x) - F_n^{(1)}(x)) + \int_{-\infty}^{+\infty} J'(\hat{F}_n(x)) (\hat{F}_n(x) - F_n(x)) dF_n^{(1)}(x) \right).$$

Since  $S_n^*(J) = V_n^*(J) + U_n(J)$  where  $\mathbf{E}(U_n(J))^2 = O(n^{-1/2})$  we prove Theorem 4.1 following the line of argument in the proof of Theorem 3.1.

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